Data structure

DEFINITION Let $\Pi = \langle C_1, \ldots, C_n \rangle$ be a partition of the set of all states. Then a factored belief space is $\langle G_1, \ldots, G_n \rangle$ where $s \subset s'$ for no $\{s, s'\} \subseteq G_i$ and $G_i \subseteq 2^{C_i}$ for all $i \in \{1, \ldots, n\}$.

Intuitively, a factored belief space is a set of belief states, partitioned to subsets corresponding to the observational classes.

The factored representation of one set S of states is $\mathcal{F}(S) = \langle \{C_1 \cap S\}, \dots, \{C_n \cap S\} \rangle$.

```
Jussi Rintanen
```

July 12, AI Planning 1/23

Data structure: combination

When we have two sets of belief states in the factored form, we may combine them and keep the result in the factored form.

DEFINITION Let $G = \langle G_1, \ldots, G_n \rangle$ and $H = \langle H_1, \ldots, H_2 \rangle$ be factored belief spaces. Define $G \oplus H$ as $\langle G_1 \sqcup H_1, \ldots, G_n \sqcup H_n \rangle$, where the operation \sqcup takes union of two sets of sets and eliminates sets that are not set-inclusion maximal. It is formally defined as $G \sqcup H = \{R \in G \cup H | R \subset K \text{ for no } K \in G \cup H\}$.

Jussi Rintanen

July 12, AI Planning 2/23

Data structure: sets of states represented by

A factored belief space $G = \langle G_1, \ldots, G_n \rangle$ can be viewed as representing the set of sets of states

 $\mathsf{flat}(G) = \{s_1 \cup \cdots \cup s_n | s_i \in G_i \text{ for all } i \in \{1, \dots, n\}\}.$

The cardinality of flat(G) is $|G_1| \cdot |G_2| \cdot \ldots \cdot |G_n|$.

DEFINITION A fbs *G* is included in fbs *H* ($G \sqsubseteq H$) if for all $S \in$ flat(*G*) there is $S' \in$ flat(*H*) such that $S \subseteq S'$. Now $S \in$ flat(*G*) if and only if $\mathcal{F}(S) \sqsubseteq G$. THEOREM Testing $G \sqsubseteq H$ for factored belief spaces *G* and *H* is polynomial time.

Data structure: inclusion

PROOF Testing $\langle G_1, \ldots, G_n \rangle \sqsubseteq \langle H_1, \ldots, H_n \rangle$ is simply by testing whether for all $i \in \{1, \ldots, n\}$ and all $s \in G_i$ there is $t \in H_i$ such that $s \subseteq t$.

Jussi Rintaner

July 12, AI Planning 4/23

```
Jussi Rintanen
```

July 12, AI Planning 3/23

Finding new belief states

```
\begin{array}{l} \textit{PROCEDURE findnew}(o,A,F,H);\\ \textit{IF } F = \langle \rangle \textit{ AND spreimg}_o(A) \not\in \textit{flat}(H) \textit{ THEN RETURN } A;\\ \textit{IF } F = \langle \rangle \textit{ THEN RETURN } \emptyset;\\ \textit{F is } \langle \{f_1, \ldots, f_m\}, F_2, \ldots, F_k\rangle \textit{ for } k \geq 1;\\ \textit{FOR } i := 1 \textit{ TO } m \textit{ DO}\\ \textit{ Z } := \textit{findnew}(o,A \cup f_i, \langle F_2, \ldots, F_k\rangle, H);\\ \textit{IF } Z \neq \emptyset \textit{ THEN RETURN } Z;\\ \textit{END};\\ \textit{RETURN } \emptyset \end{array}
```

```
Jussi Rintanen
```

July 12, AI Planning 5/23

Complexity of finding new belief states

THEOREM Testing whether $G = \langle G_1, \ldots, G_n \rangle$ contains a set S such that *spreimg*_o(S) is not in G is NP-complete. This holds even for deterministic operators o.

PROOF: Membership in NP is trivial: nondeterministically choose $s_i \in G_i$ for every $i \in \{1, ..., n\}$, compute the preimage r of $s_1 \cup \cdots \cup s_n$, verify that $r \cap C_i$ for some C_i is not in G_i .

Jussi Rintanen

July 12, AI Planning 6/23

Complexity of finding new belief states

NP-hardness by reduction from SAT. We illustrate the proof by an example. Let $T = \{A \lor B \lor C, \neg A \lor B, \neg C\}$.

Construct FBS so that *T* is satisfiable iff strong preimage of $o(x) = x_0$ is not in the FBS: clause is mapped to the set of literals not in it; satisfying valuation = a new belief state.

```
 \begin{array}{l} \langle & \{\{\widehat{A}, \widehat{B}, \widehat{C}\}, \{A, \widehat{B}, C, \widehat{C}\}, \{A, \widehat{A}, B, \widehat{B}, C\}\}, \\ & \{\{A_0\}, \{\widehat{A}_0\}\}, \\ & \{\{B_0\}, \{\widehat{B}_0\}\}, \\ & \{\{C_0\}, \{\widehat{C}_0\}\} \rangle \end{array}
```

```
Jussi Rintanen
```

July 12, AI Planning 7/23

A planning algorithm: plan(I,O,G); $H := \mathcal{F}(G)$; progress := true;

 $\begin{array}{l} \textit{WHILE progress and } I \not\subseteq S \text{ for all } S \in \textit{flat}(H) \textit{ DO} \\ \textit{progress := false;} \\ \textit{FOR EACH } o \in O \textit{ DO} \\ S := \textit{findnew}(o, \emptyset, H, H); \\ \textit{IF } S \neq \emptyset \textit{ THEN} \\ \textit{BEGIN} \\ H := H \oplus \mathcal{F}(\textit{spreimg}_o(S)); \\ \textit{progress := true;} \\ \textit{END; END; END;} \\ \textit{IF } I \subseteq S \textit{ for some } S \in \textit{flat}(H) \textit{ THEN plan found;} \end{array}$

Jussi Rintanen

July 12, AI Planning 8/23

EXPSPACE-completeness of UO planning

THEOREM 1. The problem of testing the existence of a plan for problem instances without observability is EXPSPACE-hard.

Proof idea: Simulate deterministic Turing machines with an exponential space bound.

THEOREM 2. The problem of testing the existence of a plan for problem instances without observability is in EXPSPACE.

Proof idea: Adapt the proof PSPACE-membership proof of deterministic planning to work at the level of belief states (easy!)

Jussi Rintanen

July 12, AI Planning 9/23

EXPSPACE-hardness of UO planning

Problem: exponentially many tape cells!! Representing them all requires exponentially many state variables. Reduction not polynomial time!!

Solution: Randomized test whether the plan describes an execution of the TM.

- 1. For every execution randomly choose a watched tape cell.
- Check that the plan correctly represents the watched tape cell. (Because of unobservability the plan cannot "see" what the watched tape cell is and it always has to simulate correctly.)

Jussi Rintanen

July 12, AI Planning 10/23

EXPSPACE-hardness of UO planning

Let $\langle \Sigma, Q, \delta, q_0, g \rangle$ be any deterministic Turing machine with an exponential space bound e(x).

Let σ be an input string of length n. We denote the *i*th symbol of σ by σ_i .

The Turing machine may use space e(n), and for encoding numbers from 0 to e(n) + 1 corresponding to the tape cells we need $m = \lceil \log_2(e(n) + 2) \rceil$ bits.

Jussi Rintanen

July 12, AI Planning 11/23

EXPSPACE-hardness of UO planning

The set P of state variables in the problem instance consists of

- 1. $q \in Q$ for denoting the internal states of the TM,
- 2. w_i for $i \in \{0, \dots, m-1\}$ for representing the watched tape cell $j \in \{0, \dots, e(n) + 1\}$,
- **3.** $s \in \Sigma \cup \{|, \Box\}$ for the contents of the watched tape cell,
- 4. h_i for $i \in \{0, ..., m-1\}$ for representing the position of the R/W head $j \in \{0, ..., e(n) + 1\}$.

```
Jussi Rintanen
```

EXPSPACE-hardness of UO planning

In the initial state any tape cell could be the watched one. I is the conjunction of the following formulae.

```
1. q_0
2. \neg q for all q \in Q \setminus \{q_0\}.
```

Formulae for having the contents of the watched tape cell in state variables ∑ ∪ {|, □}.

$$\begin{array}{ll} | & \leftrightarrow & (w=0) \\ \Box & \leftrightarrow & (w>n) \\ s & \leftrightarrow & \bigvee_{i \in \{1,...,n\}, \sigma_i = s} (w=i) \text{ for all } s \in \Sigma \end{array}$$

Jussi Rintanen

July 12, AI Planning 13/23

4. h = 1 for the initial position of the R/W head.

w = i, w > i denote the formulae for testing integer equality and inequality of the numbers encoded by w_0, w_1, \ldots

Later we will use effects h := h + 1 and h := h - 1 that increment and decrement the number encoded by h_0, h_1, \ldots .

Jussi Rintanen

July 12, AI Planning 14/23

EXPSPACE-hardness of UO planning

The goal is the following formula.

$$G = \bigvee \{q \in Q | g(q) = \mathsf{accept}\}$$

EXPSPACE-hardness of UO planning
For all $\langle s,q \rangle \in (\Sigma \cup \{ ,\Box\}) \times Q$ and $\langle s',q',m \rangle \in (\Sigma \cup \{ \}) \times Q \times \{L,N,R\}$ define the effect $\tau_{s,q}(s',q',m)$ as $\alpha \wedge \kappa \wedge \theta$.

 α = effect for change in current tape cell

- κ = effect for change in state of TM
- θ = effect for tape movement

Jussi Rintanen

July 12, AI Planning 16/23

Jussi Rintanen

July 12, AI Planning 15/23

EXPSPACE-hardness of UO planning, α **EXPSPACE-hardness of UO planning**, κ α describes what happens to the tape symbol under the R/W κ describes the change to the internal state of the TM. head. $\kappa = \top$ if q = q' $\alpha = \top$ if the tape cell does not change, i.e. s = s'. $\kappa = \neg q \land q'$ if $q \neq q'$ and movement $\neq R$ Otherwise. $\kappa = \neg q \land ((h < e(n)) \triangleright q')$ movement is R and $q \neq q'$ $\alpha = (h = w) \triangleright (\neg s \land s')$ The condition h < e(n) prevents reaching an accepting state if to change the watched tape cell. the space bound is violated. Jussi Rintanen Jussi Rintanen July 12, AI Planning 17/23 July 12, AI Planning

EXPSPACE-hardness of UO planning, θ

 θ describes the movement of the R/W head. Either there is movement to the left, no movement, or movement to the right. Hence 1

$$\theta = \begin{cases} h := h - 1 & \text{if } m = L \\ & \top & \text{if } m = N \\ h := h + 1 & \text{if } m = R \end{cases}$$

EXPSPACE-hardness of UO planning

Consider $\langle s,q \rangle \in \Sigma \times Q$. If $g(q) = \exists$ and $\delta(s,q) = \langle s',q',m \rangle$, then define

$$o_{s,q} = \langle ((h \neq w) \lor s) \land q \land (h \le e(n)), \tau_{s,q}(s',q',m) \rangle$$

The condition $(h \neq w) \lor s$ is the key to the EXPSPACE-hardness proof: If the plan tries to cheat here, then the operator is not applicable on some execution, and the plan is not a valid plan.

Jussi Rintanen

July 12, AI Planning 20/23

18/23

Jussi Rintanen

July 12, AI Planning 19/23



2-EXPTIME-hardness of PO planning

- Branches in plans make it possible to simulate alternating TMs that have a computation tree with AND and OR nodes.
- With partial observability we can extend the EXPSPACE TM simulation to AEXPSPACE = 2-EXPTIME ($O(2^{2^n})$ time) simulation.

AND node = nondeterministic operator corresponds to a set of transitions OR node = a set of deterministic operators corresponds to a set of transitions (one operator is chosen)

Jussi Rintanen

July 12, AI Planning 23/23